Optimal control of path-dependent McKean-Vlasov SDEs in infinite dimension

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## Motivation and examples

### 2 Our Mc Kean - Vlasov type setting

## 3 Our results

- Law invariance for the value function
- Dynamic programming
- Chain rule
- HJB PDE
- Comparison and uniqueness
- 4 Further ongoing research



- Optimal control of McKean-Vlasov type SDEs is a recent topic which arises in a natural way in problems where the dynamics of the state equation depends on the state/control law.
- Typical "planner's problem" in economics with many players, as opposed to "agent's problem" (→ Mean Field Game).
- Various papers recently have studied the case when state equation is finite dimensional (see e.g. the book [Carmona-Delarue] and various papers) and possibly path-dependent [Wu-Zhang].

- Up to our knowledge, no paper studies the infinite dimensional case. However such case arises naturally in applications.
- We then aim to develop the theory in this case trying also to clarify some issues left in previous papers. This paper is the first step.

# Example 1: spatial economic growth models

Spatial growth models (see e.g. [Boucekkine-Camacho-Fabbri '16] [Gozzi-Leocata '21]) lead to optimal control in an infinite dimensional space *H*.

Typical issue in such problems: the productivity depends on the average of the capital distribution (see e.g. [Turnovsky '06]).  $\Rightarrow$  state equation for the capital trajectory  $k(\cdot)$  as

$$dk(t) = [Ak(t) + a(\mathbb{E}k(t))k(t) - \delta k(t) - c(t)]dt + \sigma(k(t))dB(t).$$

Here A is the laplace operator,  $c(\cdot)$  is the control (consumption rate),  $a(\cdot)$ ,  $\delta$ ,  $\sigma(\cdot)$  are given data.

Also need to include in such type of models, delay/path-dependent features like time-to build or vintage capital.

# Example 2: Lifecycle portfolio with "sticky" wages

In such problems (see e.g. [Djeiche-Gozzi-Zanco-Zanella '20]), labor income " $y(\cdot)$ " (one of the state equations of the optimal portfolio problem) described by one-dimensional delay SDEs of McKean-Vlasov type as follows

$$dy(t) = \left[b_0(\mathbb{P}_{y(t)}) + \int_{-d}^0 y(t+\xi)\phi(\xi)\,d\xi\right]dt + \sigma y(t)\,dZ(t).$$

(here  $\phi \in L^2(-d, 0; \mathbb{R})$  is a given datum and Z is a one-dimensional Brownian motion). Such equations can be rephrased as SDEs in the Hilbert space  $\mathbb{R} \times L^2(-d, 0; \mathbb{R})$  and the resulting dynamics is a Mc Kean - Vlasov SDEs in infinite dimension.



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# Basic and probabilistic setting

- The state space *H* and the control space *K* are real separable Hilbert spaces. The control set *U* is a Polish space or a Borel subset of it. The horizon *T* is finite.
- (Ω, ℱ, ℙ, (ℱ<sup>B</sup><sub>s</sub>)<sub>s∈[0, T]</sub>, B) is a reference probability space i.e. a complete probability space with a cylindrical Brownian motion with values in K and (ℱ<sup>B</sup><sub>s</sub>)<sub>s∈[0, T]</sub> is the augmented filtration generated by B.
- There exists a sub- $\sigma$ -algebra  $\mathscr{G}$  of  $\mathscr{F}$ , independent of  $\mathscr{F}^B_{\infty}$  and admitting a  $\mathscr{G}$ -measurable r.v.  $U_{\mathscr{G}}$  with uniform distribution in [0,1].
- We call  $\mathbb{F} := (\mathscr{F}_s)_{s \in [0,T]}$  with  $\mathscr{F}_s = \mathscr{G} \lor \mathscr{F}_s^B$ ,

## State equation

 $X(\cdot)$  is the state while  $\alpha(\cdot)$  is the control:

$$dX(s) = AX(s) + b(s, X_s, \mathbb{P}_{X_s}, \alpha(s), \mathbb{P}_{\alpha(s)})ds + \sigma(s, X_s, \mathbb{P}_{X_s}, \alpha(s), \mathbb{P}_{\alpha(s)})dB(s), \quad s > t$$

and  $X(s) = \xi(s)$  for  $0 \le s \le t$ . Here:

- A is a linear unbounded operator (generating a strongly continuous semigroup);
- b and σ are progressive;
- subscript s denotes the history of the process up to time s.

#### Assume

- b and σ are bounded and satisfy a Lipschitz condition wrt state and measure (in the Wasserstein space (𝒫<sub>2</sub>, 𝒱<sub>2</sub>));
- the initial datum  $\xi$  belongs to  $S_2(\mathbb{F})$  which is the set of *H*-valued continuous  $\mathbb{F}$ -progressively measurable processes  $\xi$  such that  $\|\xi\|_{S_2} := \left(\mathbb{E}[\sup_{t \in [0,T]} |\xi(t)|_H^2]\right)^{1/2} < +\infty.$

Then, the controlled process X is well defined in  $S_2(\mathbb{F})$ .

# Objective function

We aim to maximize the functional

$$J(t,\xi;\alpha) = \mathbb{E}\left[g\left(X_T^{t,\xi,\alpha}, \mathbb{P}_{X_T^{t,\xi,\alpha}}\right) + \int_t^T f\left(s, X_s^{t,\xi,\alpha}, \mathbb{P}_{X_s^{t,\xi,\alpha}}, \alpha(s), \mathbb{P}_{\alpha(s)}\right) ds\right]$$

Here:

- f and g are progressive, continuous, locally bounded and locally uniformly continuous in state and measure (uniformly wrt the other variables) and with quadratic growth in state.
- α ∈ 𝔐 where 𝔐 is the space of 𝔽-progressively measurable processes [0, 𝒯] × Ω → 𝕖.
- $X^{t,\xi,\alpha}$  is the solution of the state equation with initial data  $(t,\xi) \in [0,T] \times \mathbf{S}_2(\mathbb{F})$  and control  $\alpha \in \mathscr{U}$ .

# Value function

The value function  $V : [0, T] \times \mathbf{S}_2 \to \mathbb{R}$  is defined by

$$V(t,\xi) = \sup_{\alpha(\cdot)\in\mathscr{U}} J(t,\xi;\alpha), \qquad (t,\xi)\in[0,T]\times\mathbf{S}_2.$$

On the line of the previous papers on the finite dimensional case, in particular [Wu-Zhang,'20], we expect that the value function only depends on the law of the initial datum

We also expect that its equivalent v on measures solves, in a suitable viscosity sense, the associated HJB equation on the space of measures  $\mathscr{P}_2(C([0, T]; H))$ .

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# Summary of the results

In the paper we prove the following.

- Law invariance of V, i.e. that V(t, ξ) only depends on the law of ξ up to time t.
- Dynamic Programming Principle.
- Ito formula in the context of pathwise derivatives in the spirit of Dupire and Wu-Zhang definition.
- Viscosity property of V for the HJB equation.
- Uniqueness in a specific case.

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# Comments of results

Obtained results are those expected (as the community becomes more and more used to this mean field framework).

However, proofs involve a lot of technicalities (the paper is 54 pages long!).

Rest of the talk : try to stay reasonable (focus on the main steps)

Some comments on further ongoing work will also be given.

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# Law invariance of V

- The goal is to prove that, if  $\xi$  and  $\eta$  belong to  $S_2(\mathbb{F})$ , with  $\mathbb{P}_{\xi} = \mathbb{P}_{\eta}$ , then  $V(t,\xi) = V(t,\eta)$
- This was proved, in the finite dimensional case, in [Cosso-Pham,'18]. However their proof uses a result from [Aliprantis-Border,'06], Corollary 18.23, which is not correct as it is. Hence their proof does not work.
- Our proof is based on the fact that one can find, for every  $\xi$ ,  $\eta$  above, two r.v.  $U_{\xi}$  and  $U_{\eta}$ , with uniform law on [0,1], such that  $\xi$  and  $U_{\xi}$  (and also  $\eta$  and  $U_{\eta}$ ) are independent. A
- We also provide and example where this does not apply.

Define v by  $v(t,\mu) = V(t,\xi)$  for any  $\xi$  with  $\mathbb{P}_{\xi} = \mu$  (lift-inverse).

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# DPP for V

#### Theorem

Under Assumption on the coefficients  $b,\sigma,f,g$ , the lifted value function V satisfies the **dynamic programming principle**:

$$V(t,\xi) = \sup_{\alpha \in \mathcal{U}} \left\{ \mathbb{E} \left[ \int_{t}^{s} f_r(X^{t,\xi,\alpha}, \mathbb{P}_{X^{t,\xi,\alpha},\alpha}, \alpha_r, \mathbb{P}_{\alpha_r}) dr \right] + V(s, X^{t,\xi,\alpha}) \right\}$$

for every  $t, s \in [0, T]$ , with  $t \leq s$ , and  $\xi \in S_2(\mathbb{F})$ .

Rq: no measurable selection issue as the function V depends on the whole r.v.  $\xi.$ 

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# DPP for the lift inverse v

## Corollary

Under previous assumptions on the coefficients  $b,\sigma,f,g$ , the value function v satisfies the **dynamic programming principle**:

$$v(t,\mu) = \sup_{\alpha \in \mathscr{U}} \left\{ \mathbb{E} \left[ \int_t^s f_r(X^{t,\xi,\alpha}, \mathbb{P}_{X^{t,\xi,\alpha}, \alpha_r}, \mathbb{P}_{\alpha_r}) dr \right] + v(s, \mathbb{P}_{X^{t,\xi,\alpha}}) \right\},$$

for every  $t, s \in [0, T]$ , with  $t \le s$ ,  $\mu \in \mathscr{P}_2(C([0, T]; H))$  and  $\xi \in \mathbf{S}_2(\mathbb{F})$  with  $\mathbb{P}_{\xi} = \mu$ .

Since V non-anticipative, namely  $V(t,\xi) = V(t,\xi_{\cdot,t})$ , for every  $(t,\xi) \in [0,T] \times S_2(\mathbb{F})$ , v satisfies

$$v(t,\mu)=v(t,\mu_{[0,t]}),$$

where  $\mu_{[0,t]} = \mu \circ ((x_s)_{s \in [0,T]} \mapsto (x_{s \wedge t})_{s \in [0,T]})^{-1}$ . Idris Kharroubi Optimal control of path-dependent McKean-Vlasov SDEs in

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We need to define derivatives in this path-dependent framework. That is

- time derivative,
- measure derivative,
- path × measure derivative.

Use a definition via the lift function. Denote by  $\Phi$  the lift of  $\varphi$ :

$$\Phi(t,\xi) = \varphi(t,\mathbb{P}_{\xi}), \qquad \forall (t,\xi).$$

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## Time derivative

Define

$$\mathscr{H} = [0, T] \times \mathscr{P}_2(C([0, T]; H)).$$

 $\varphi$  is pathwise differentiable in time at  $(t,\mu) \in \mathcal{H}$ , if the following limit

$$\partial_t \varphi(t,\mu) = \lim_{\delta \to 0^+} \frac{\varphi(t+\delta,\mu_{[0,t]}) - \varphi(t,\mu)}{\delta}$$

exists and is finite.

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## Measure derivative

 $\Phi$  is **pathwise differentiable in space** at  $(t,\xi)$  if there exists  $D\Phi(t,\xi) \in L^2(\Omega; H)$  such that

$$\lim_{Y\to 0} \frac{\left|\Phi(t,\xi+Y\mathbf{1}_{[t,T]}) - \Phi(t,\xi) - \mathbb{E}[\langle D\Phi(t,\xi), Y\rangle_H]\right|}{|Y|_{L^2(\Omega;H)}} = 0.$$

Then, there exists a measurable function  $\partial_{\mu}\varphi(t,\mu)(.)$  such that

 $D\Phi(t,\xi) = \partial_{\mu}\varphi(t,\mu)(\xi)$  P-a.s.

The function  $\partial_{\mu}\varphi(t,\mu)$  is the measure derivative of  $\varphi$  at  $(t,\mu) \in \mathcal{H}$ .

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# Path × measure derivative

We say that  $\varphi$  is **pathwise differentiable in measure and space** at  $(t, \mu, x)$  if there exists an operator  $\partial_x \partial_\mu \varphi(t, \mu)(x) \in \mathscr{L}(H)$  such that

$$\lim_{h\to 0} \frac{\left|\partial_{\mu}\varphi(t,\mu)(x+h\mathbf{1}_{[t,T]})-\partial_{\mu}\varphi(t,\mu)(x)-\partial_{x}\partial_{\mu}\varphi(t,\mu)(x)h\right|_{H}}{|h|_{H}}=0.$$

We refer to  $\partial_x \partial_\mu \varphi(t,\mu)(x)$  as the second-order pathwise derivative in measure and space of  $\varphi$  at  $(t,\mu,x)$ .

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$$C_b^{1,2}(\mathscr{H})$$
:  $\varphi$  s.t.  $\varphi$ ,  $\partial_t \varphi$ ,  $\partial_\mu \varphi$ ,  $\partial_x \partial_\mu \varphi$  are cont. and bounded.  
Fix  $t \in [0, T]$  and  $\xi \in \mathbf{S}_2(\mathbb{F})$ .

Let  $F: [0, T] \times \Omega \to H$ ,  $G: [0, T] \times \Omega \to \mathscr{L}_2(K; H)$   $\mathbb{F}$ -progressive, such that

$$\int_0^T \mathbb{E}[|F_s|_H^2] \, ds < \infty, \qquad \qquad \int_0^T \mathbb{E}[\operatorname{Tr}(G_s G_s^*)] \, ds < \infty.$$

Consider the process  $X = (X_s)_{s \in [0, T]}$  given by

$$X_s = \xi_{s \wedge t} + \int_t^{s \vee t} F_r \, dr + \int_t^{s \vee t} G_r \, dB_r, \qquad \forall s \in [0, T].$$

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#### Theorem

If  $\varphi : \mathcal{H} \to \mathbb{R}$  is in  $C_b^{1,2}(\mathcal{H})$ , then the following Itô formula holds:

$$\varphi(s, \mathbb{P}_{X_{\cdot, \wedge s}}) = \varphi(t, \mathbb{P}_{\xi_{\cdot, \wedge t}}) + \int_{t}^{s} \partial_{t} \varphi(r, \mathbb{P}_{X_{\cdot, \wedge r}}) dr$$
$$+ \int_{t}^{s} \mathbb{E}[\langle F_{r}, \partial_{\mu} \varphi(r, \mathbb{P}_{X_{\cdot, \wedge r}})(X_{\cdot, \wedge r}) \rangle_{H}] dr$$
$$+ \frac{1}{2} \int_{t}^{s} \mathbb{E}[\operatorname{Tr}(G_{r} G_{r}^{*} \partial_{X} \partial_{\mu} \varphi(r, \mathbb{P}_{X_{\cdot, \wedge r}})(X_{\cdot, \wedge r}))] dr,$$

for every  $s \in [t, T]$ .

# HJB equation

Let  $\mathcal{M}_t := \{\mathfrak{a} \colon \Omega \to U \colon \mathfrak{a} \text{ is } \mathscr{F}_t\text{-measurable}\}$ . HJB PDE writes

$$0 = \partial_{t}w(t,\mu) + \mathbb{E}\langle\xi_{t},A^{*}\partial_{\mu}w(t,\mu)(\xi)\rangle_{H} + \sup_{\mathfrak{a}\in\mathcal{M}_{t}} \Big\{ \mathbb{E}[f_{t}(\xi,\mu,\mathfrak{a},\mathbb{P}_{\mathfrak{a}}) + \langle b_{t}(\xi,\mu,\mathfrak{a},\mathbb{P}_{\mathfrak{a}}),\partial_{\mu}w(t,\mu)(\xi)\rangle_{H}] + \frac{1}{2}\mathbb{E}\Big[\mathsf{Tr}\Big(\sigma_{t}(\xi,\mu,\mathfrak{a},\mathbb{P}_{\mathfrak{a}})\sigma_{t}^{*}(\xi,\mu,\mathfrak{a},\mathbb{P}_{\mathfrak{a}})\partial_{x}\partial_{\mu}w(t,\mu)(\xi)\Big)\Big]\Big\},$$

for  $(t,\mu) \in \mathscr{H}$ , t < T,  $\xi \in S_2(\mathscr{G})$  s.t.  $\mathbb{P}_{\xi} = \mu$ ,

with terminal condition

$$w(T,\mu) = \mathbb{E}[g(\xi,\mu)]$$

for  $\mu \in \mathscr{P}_2(C([0, T]; H)), \xi \in \mathbf{S}_2(\mathscr{G})$  s.t.  $\mathbb{P}_{\xi} = \mu$ .

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# **Regular solutions**

We say that a function w: ℋ→ ℝ belongs to the space C<sup>1,2</sup><sub>b,A\*</sub>(ℋ) if it satisfies the following regularity assumptions:
(i) w: ℋ→ ℝ belongs to C<sup>1,2</sup><sub>b</sub>(ℋ);
(ii) for all (t,μ,ξ) ∈ H×S<sub>2</sub>(ℙ), ∂<sub>μ</sub>φ(t,μ)(ξ) ∈ L<sup>2</sup>(Ω; D(A\*)) and the map

$$\mathscr{H} \times \mathbf{S}_2(\mathbb{F}) \longrightarrow L^2(\Omega; H), \qquad (t, \mu, \xi) \longmapsto A^* \varphi(t, \mu)(\xi)$$

is continuous and bounded.

• We say that a function  $w: \mathscr{H} \to \mathbb{R}$  is a classical solution to the HJB equation if it belongs to the space  $C_{b,A^*}^{1,2}(\mathscr{H})$  and satisfies the HJB PDE.

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# **Regular solutions**

#### Theorem

Suppose that previous assumptions on  $b, \sigma, f, g$  hold and that  $b, \sigma, f$  are uniformly continuous in t, uniformly with respect to the other variables. Assume that the value function v belongs to the space  $C_{b,A^*}^{1,2}(\mathcal{H})$ . Then v is a classical solution of the HJB PDE.

Consequence of Itô formula and DPP applied to the function v.

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# viscosity solutions

Definition of viscosity solutions same as classical one except that it uses test functions  $\varphi \in C_{b,A^*}^{1,2}(\mathcal{H})$ .

#### Theorem

Under previous assumptions on  $b,\sigma,f$  and g, the value function v is a viscosity solution to the HJB PDE.

The proof follows exactly the same lines as for the regular case, simply replacing v with a test function  $\varphi$ .

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Comparison: back to  $\mathbb{R}^d$  and no path dependency

Second-order HJB equations in the Wasserstein space is emerging and still at an early stage.

[Burzoni et al. 20] study a special class of HJB equations:  $b, \sigma, f, g$  do not depend on x, and control is deterministic.

[Wu & Zhang 20] use a notion of viscosity solution with conditions formulated on compact subsets of the Wasserstein space.

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# Comparison

## Theorem (Comparison)

Suppose that previous assumptions hold,  $\sigma$  does not dependent on  $\mu$  (the law) and is  $C_b^2$ . Let  $u_1, u_2: [0, T] \times \mathscr{P}_2(\mathbb{R}^d) \to \mathbb{R}$  be continuous bounded functions. Suppose that  $u_1$  (resp.  $u_2$ ) is a viscosity subsolution (resp. supersolution) of HJB equation . Then

 $u_1 \leq u_2, \quad on [0, T] \times \mathscr{P}_2(\mathbb{R}^d).$ 

## Corollary (Uniqueness)

Under previous assumptions, v is the unique bounded and continuous viscosity solution of HJB equation.

Not possible to use classical Ishii's Lemma based approach.

Back to the original proof: prove that  $u_1 \le v \le u_2$ .

lssues

- Need a regular version/approximation of  $v \Rightarrow$  replace v by its *n*-agent/particle (common) optimization value  $v_n$ .
- No local compactness ⇒ use a smooth variational principle of Borwein-Preiss type with perturbation/gauge function constructed via a partition of the space as in [Dereich et al.13] and a convolution with Gaussians.

# Outline

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- Cases where the solutions of HJB equation are regular and the optimal feedback control can be found.
- Mean Field Games in infinite dimension (with S. Federico and M. Rosestolato).
- Applications to specific problems.

## Thank you for your attention